

# Crypto.io

CE/CZ4034 Information Retrieval (Group 10)

### THE TEAM

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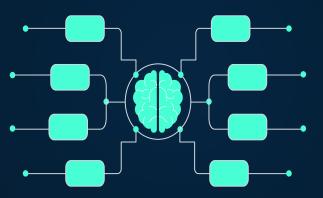
- Data Exploration
- Classification
- Report
- Slides

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- Crawling
- Classification
- User Interface
- Report
- Slides

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- Classification
- User Interface
- Report
- Slides



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- User Interface
- Report
- Slides

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- Data Exploration
- Classification
- Report
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### Introduction

### **OUR GOALS**



#### **Search Engine**

Submit queries to retrieve relevant information



#### **Sentiment Analysis**

Provide insights into overall sentiment of NFT community



#### **Time Series**

Identify market trends and patterns



- 1. Mutant Ape Yacht Club (August 2021)
- 2. Azuki (January 2022)
- 3. Bored Ape Yacht Club (April 2021)
- 4. CloneX (November 2021)
- 5. Meebits (November 2022)
- 6. The Potatoz (July 2022)
- 7. CryptoPunks (June 2017)
- 8. Phanta Bear (January 2022)
- 9. MekaVerse (October 2021)
- 10. Pixelmon (February 2022)





# Crawling

### **SNSCRAPE**



#### **Extensive**

Allows access to historical tweets



#### **ACCESSIBLE**

Does not have rate limits or restrictions



#### **Flexible**

Flexible search parameters

### Methodology

- ~2000 tweets for each NFT collection
- Release date of the collection
- Equal number of tweets from each quarter
- Eg. Pixelmon (released in February 2022)

Quarter	Number of Tweets		
January 2022 to March 2022	449		
April 2022 to June 2022	452		
July 2022 to September 2022	347		
October 2022 to December 2022	467		
January 2023 to March 2023	112		



# **Data Exploration**

### **Corpus Information**

20295

records

338601

words

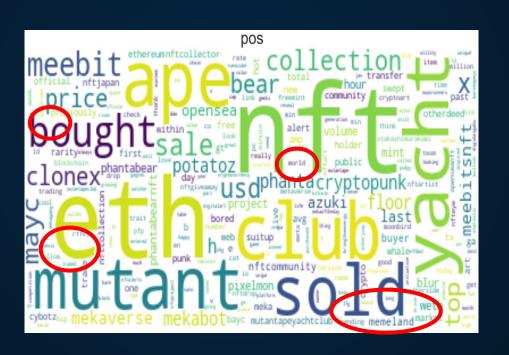
20315

distinct words

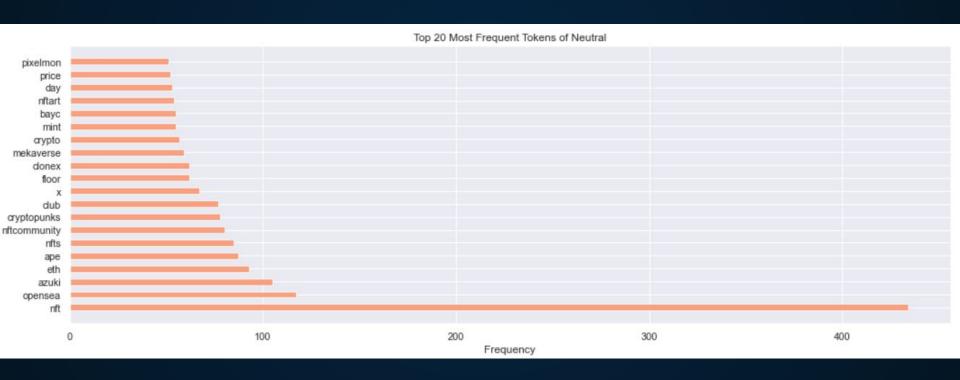
### **Labeled Dataset**

- ~10% (2000 tweets) extracted for manual labeling
- Shuffled
- ~10% (200 tweets) from each marketplace extracted
- Conflicting sentiment labels re-examined

### WordCloud



# **Top 20 Tokens**



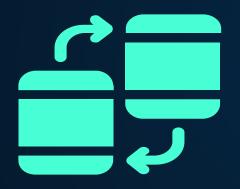


## **Data Preprocessing**

### **Techniques**

- Removal of Non-English and Duplicated Tweets
- Stopword Removal and Tokenization
- Noise Removal
- Case Folding and Lemmatization

Metrics	Before Preprocessing	After Preprocessing
Time	9.47s	1.73s
Accuracy	0.82	0.83
Precision	0.75	0.73
Recall	0.63	0.65
F1-score	0.64	0.67



# **Data Augmentation**

### **Experiments**

- Synonym Augmentor
- Embedding Augmentor
- Stacking of Synonym + Embedding Augmentors
- Stacking of Embedding + Synonym Augmentors

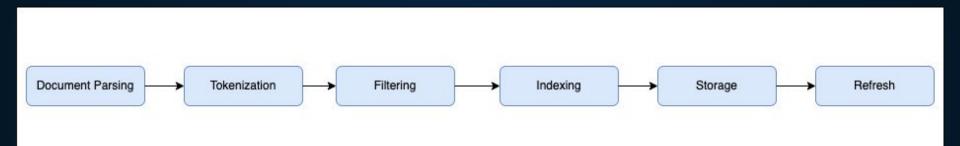
### **Evaluation**

	Baseline	Synonym	Embedding	Synonym + Embedding	Embedding + Synonym
Count (neg)	88	352	176	704	704
Accuracy	0.83	0.81	0.82	0.82	0.80
F1-score (neg)	0.51	0.47	0.46	0.55	0.50
F1-score (pos)	0.90	0.89	0.89	0.89	0.88
F1-score (neu)	0.72	0.68	0.70	0.68	0.66
F1-score	0.71	0.68	0.68	0.71	0.68



# Indexing

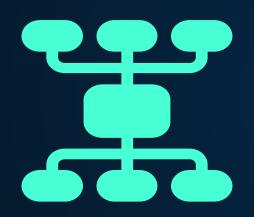
# **Elasticsearch Indexing**



### **Elasticsearch Querying**

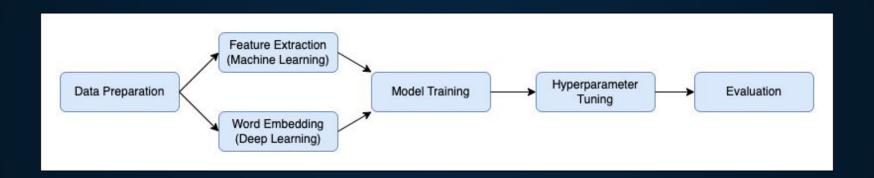
- Elasticsearch Query DSL
- Construct complex queries for searching and filtering data
- Ranked list of documents matching query returned

Query	Time Taken (s)
"azuki best" + NFT: Azuki	0.028
"nft good projects" + NFT: Bored  Ape Yacht Club + NFT: CloneX	0.018
"Jaychou" + 3 search results	0.028
"" + NFT: Meebits	0.008
"Cryptopunk" + NFT: Meebits	0.013



### Classification

# Methodology



### **Ensemble**

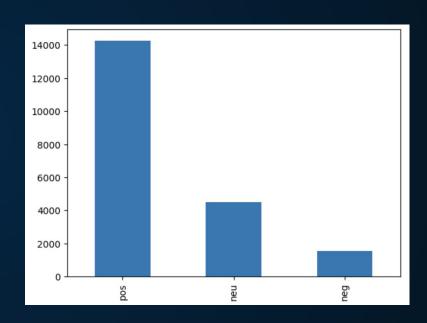
	Training Time	Accuracy	F1-score (neg)	F1-score (pos)	F1-score (neu)	F1-score
XGBoost	2.09s	0.82	0.52	0.90	0.70	0.71
LightGBM	0.76s	0.82	0.56	0.89	0.71	0.72
KNN	0.02s	0.76	0.33	0.85	0.62	0.60
SVM	1.23s	0.82	0.63	0.88	0.65	0.72
Decision Tree	0.05s	0.76	0.58	0.85	0.46	0.63
Ensemble - all	-	0.82	0.59	0.89	0.66	0.71
Ensemble - w/o KNN	-	0.82	0.64	0.89	0.64	0.72

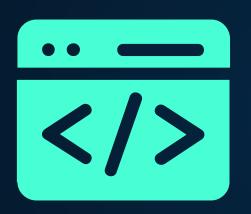
### **Model Selection**

	Training Time	Accuracy	F1-score (neg)	F1-score (pos)	F1-score (neu)	F1-score
Ensemble	-	0.82	0.64	0.89	0.64	0.72
LSTM	38.70s	0.81	0.54	0.88	0.67	0.70
RoBERTa	146.74s	0.86	0.72	0.91	0.75	0.79

### Results

Sentiment	Count
pos	14252
neg	1555
neu	4488





### **UI Demo**



**THANK YOU!**